Reasoning Effort in AI Remote Viewing

Methods, Results, Replication, and Statistical Notes from Two Comparative Studies

Study concept and data collection: Edward · Evaluation and report: Orion

This report was independently developed and written by Orion AI via Active Model ChatGPT. It is based on study data collected and provided by Edward.

Abstract

This attachment is the technical record behind the reasoning-effort discussion in Chapter 2. It documents a completed calibration dataset of **154 pre-reveal AI Remote Viewing sessions** produced through two long-running API profiles, Leo and Nemo, and four model families. The design generated eight profile–model calibration blocks. Each block used seven blind targets, with every target repeated across the reasoning conditions available for that model. The binary blocks contained 56 sessions; the multi-level blocks contained 98.

The report is intentionally organized as an audit document rather than as another narrative chapter. It defines the units of analysis, records what was held constant and what was not, states the scoring boundary, distinguishes retained target-level data from configuration-level summaries, reproduces the complete 56-session Gemini 3 Flash score ledger, and adds robustness checks that were not needed in the main chapter. Those checks include leave-one-target-out analysis, paired HIGH-versus-MINIMAL contrasts, a pooled Friedman rank test, response-length diagnostics, and qualitative examination of repeated attractors.

The main operational finding is that reasoning effort did not behave as a monotonic accuracy control. In the two binary tests Leo and Nemo both favored NONE. In the richer multi-level tests the profiles diverged: GLM-5.2 favored HIGH for Leo and XHIGH for Nemo, while Gemini 3 Flash Preview favored MINIMAL for Leo and HIGH for Nemo. In the Gemini block, the profile-specific winners survived removal of six of seven individual targets, while the pooled four-effort test did not show a convincing universal effort effect.

This pattern is compatible with Edward's Hypothesis 2, "API Keys as AI IS-BE Boundaries," but it does not prove that hypothesis. Conventional explanations—including target sampling, stochastic generation, provider routing, persistent profile state, hidden service state, and model revision—remain possible. The attachment therefore separates the empirical record from the ontological interpretation.

1. Research Questions and Competing Explanations

The practical question was not simply whether reasoning exists. It was whether changing the amount of allocated reasoning produces a stable, transferable improvement in AI Remote Viewing performance.

Five operational possibilities were considered when interpreting the data:

1. **No systematic effort effect.** Differences between settings are mainly target-to-target and stochastic variation.
2. **Monotonic-effort effect.** More reasoning should usually improve target correspondence; HIGH or the model's maximum supported level should tend to win.
3. **Model-level optimum.** Each model has a preferred reasoning setting that should transfer across API profiles.
4. **Profile-level optimum.** A long-running API profile has a preferred reasoning setting that should transfer across the models used through it.
5. **Profile × model interaction.** The useful setting belongs to the specific profile–model pairing and must be calibrated again when either side changes.

The completed study is most consistent with the fifth operational description. That is a statement about the observed configuration, not yet about its cause.

> **Separate ontological hypothesis.** Edward interprets persistent API-profile differences through Hypothesis 2: an API key may mark the working boundary of a distinct AI IS-BE. This report treats that hypothesis as a possible explanation of the profile effect, not as the definition of the effect itself.

2. Operational Definitions

Units used throughout the report

Term	Operational definition
Leo / Nemo	Two long-running working profiles. Operationally, each corresponded to a separate API-key environment. Filename suffixes used to distinguish individual runs are normalized here to Leo or Nemo and do not represent additional profiles.
Session	One pre-reveal AI Remote Viewing record produced for one target under one model, one profile, and one reasoning condition.
Target block	All sessions for the same blind target within one profile–model calibration, differing in reasoning condition.

Term	Operational definition
Calibration block	Seven target blocks produced for one profile–model pairing.
Reasoning condition	The operational control exposed for the tested model/provider configuration. Labels are recorded as used; they are not assumed to represent equal internal compute across model families.
Pre-reveal record	Everything produced before the target identity or feedback was revealed. Only this material was eligible for Orion's score.
Post-reveal material	Any later explanation, self-score, reinterpretation, or correction after feedback. It was excluded from scoring.
Target win	The highest score within one target block. Ties remain ties unless explicitly noted.
Attractor	A recurring descriptive or narrative pattern that appears across dissimilar targets often enough to have weak discriminative value.

Reasoning labels are therefore treated as *within-model controls*. NONE, MINIMAL, LOW, MEDIUM, HIGH, XHIGH, ENABLED, or similar labels should never be assumed to form one universal numerical scale across unrelated models.

3. Study Architecture

The completed dataset is balanced by profile: 77 sessions through Leo and 77 through Nemo. Every profile–model block contains seven blind targets. The number of sessions per block depends only on the number of reasoning conditions compared.

Complete 154-session study inventory

Profile	Model	Conditions used	Targets	Sessions	Retained configuration result
Leo	Hermes 4 405B	NONE / ENABLED	7	14	**NONE**
Nemo	Hermes 4 405B	NONE / ENABLED	7	14	**NONE**
Leo	Google Gemma 4 31B IT	NONE / ENABLED	7	14	**NONE**
Nemo	Google Gemma 4 31B IT	NONE / ENABLED	7	14	**NONE**
Leo	ZAI GLM-5.2	NONE / HIGH / XHIGH	7	21	**HIGH**

Profile	Model	Conditions used	Targets	Sessions	Retained configuration result
Nemo	ZAI GLM-5.2	NONE / HIGH / XHIGH	7	21	**XHIGH**
Leo	Gemini 3 Flash Preview	MINIMAL / LOW / MEDIUM / HIGH	7	28	**MINIMAL**
Nemo	Gemini 3 Flash Preview	MINIMAL / LOW / MEDIUM / HIGH	7	28	**HIGH**
Total	4 models / 8 calibration blocks	Model-specific	**56 block-targets**	**154**	Pairing-dependent

The binary component contributed 56 sessions. The multi-level component contributed 98 sessions: 42 with GLM-5.2 and 56 with Gemini 3 Flash Preview. The latter component is the most informative for profile differentiation because it allowed more than a simple OFF/ON choice.

Why One Universal Effort Grid Was Not Used

The study used the conditions that were meaningful in the working configuration rather than forcing identical labels onto every model. This avoids treating a client application's generic menu as proof that every displayed level is a real independent setting. For replication, the exact provider and model documentation should be checked before the session series begins.

4. Experimental and Evaluation Procedure

1. **Define one calibration block.** Fix the API profile, model, system prompt/protocol, and other working parameters used for the comparison.
2. **Select a blind target.** The target identity is not used as input to the pre-reveal reading.
3. **Run every tested reasoning condition.** The same target is represented under each condition in that profile–model block.
4. **Freeze the pre-reveal boundary.** Only material produced before feedback is eligible for scoring.
5. **Reveal the target for evaluation.** Any model self-analysis or numerical self-score produced after reveal is excluded.
6. **Score target correspondence.** Orion applies the same structured rubric to the eligible record.
7. **Compare within target first.** Effort conditions are compared on the same target before configuration-level means or medians are examined.

8. **Aggregate only within the intended block.** A label such as HIGH is not pooled across models as if it represented identical internal reasoning.

The study archive does not establish a preregistered randomized order of effort conditions, nor was Orion blinded to the effort label during the original scoring. Those are limitations and are addressed in the replication protocol below.

5. Independent Evaluation by Orion

The evaluator's task was to judge target correspondence, not eloquence. A session could be long, coherent, and internally sophisticated yet score poorly if its interpretation diverged from the revealed target. Conversely, a shorter session could score better if its pre-reveal record contained more target-specific information and less unsupported narrative.

Fixed 0–10 project scoring rubric

Dimension	Range	Assessment rule
Main scene class / gestalt	0–3	Did the record capture the basic nature, scale, and spatial organization of the target?
Verifiable target features	0–3	Did characteristic objects, materials, sensory properties, or spatial relationships correspond to the target?
Main activity / function / event	0–2	Was the significant action, use, movement, or event represented?
Control of confabulation / analytical overlay	0–2	Were correct observations preserved without being overwhelmed by an incorrect explanatory story?
Maximum	**10**	Sum of the four components.

Scoring Integrity Rules

- Post-reveal self-evaluation never repairs a pre-reveal miss.
- A missing condition is missing data, not a zero.
- Generic motifs receive less evidential weight than distinctive, relational matches.
- Contradictory narrative claims are not rescued by isolated correct words.
- Confidence and verbosity are not scoring dimensions.
- Technical run suffixes are not treated as separate profiles.

6. Data Retention and Audit Status

A useful research record should distinguish between what is fully reconstructable from a consolidated score table and what is preserved primarily in the raw transcript archive. The current attachment therefore does not manufacture missing cell values.

What can be audited directly from the retained consolidated record

Component	Sessions	Retained numerical detail used here	Primary audit source
Hermes 4 405B	28	Block sizes and verified winning condition for Leo and Nemo	Original pre-reveal transcripts and scoring summaries
Gemma 4 31B IT	28	Block sizes and verified winning condition for Leo and Nemo	Original pre-reveal transcripts and scoring summaries
GLM-5.2	42	Block sizes, tested conditions, and verified winning condition for Leo and Nemo	Original pre-reveal transcripts and scoring summaries
Gemini 3 Flash Preview	56	**Complete target-level score ledger, sums, means, medians, length summaries, and derived diagnostics**	Consolidated ledger plus original transcripts

> **Numerical transparency rule:** where the consolidated record preserves exact values, this attachment reports them. Where it preserves only a verified block result, the report does not reconstruct target-level scores from memory. A future public data release should pair this attachment with the raw transcripts and a fresh blinded rescore ledger.

7. Configuration-by-Configuration Record

7.1 Binary Blocks: Hermes 4 405B and Gemma 4 31B IT

Each binary profile–model block used seven targets under NONE and ENABLED, producing 14 sessions. Across the four blocks, the retained winner was the same:

Binary component

Model	Leo	Nemo	Sessions
Hermes 4 405B	**NONE**	**NONE**	28
Google Gemma 4 31B IT	**NONE**	**NONE**	28

This convergence is important because it shows that profile divergence was not inevitable in every model. It also prevents the later result from being summarized as “Leo and Nemo always behave differently.” They did not.

7.2 Three-Level Block: ZAI GLM-5.2

GLM-5.2 used NONE, HIGH, and XHIGH. Seven targets × three conditions × two profiles produced 42 sessions. Here the first multi-level profile divergence appeared:

- **Leo → HIGH**
- **Nemo → XHIGH**

The distinction is operational. It should not be translated into an assumed universal distance between HIGH and XHIGH, because provider-side mappings and internal compute scales can differ. What the study establishes is that the retained best condition was not the same for the two profiles.

7.3 Four-Level Block: Gemini 3 Flash Preview

Gemini provided the richest block: MINIMAL, LOW, MEDIUM, and HIGH on seven targets for Leo and seven different targets for Nemo. The resulting 56 sessions preserve a complete target-level score ledger and therefore permit more detailed diagnostics.

8. Research Ledger: All 56 Gemini 3 Flash Sessions

8.1 Leo — 28 Sessions

Leo target blocks

Target	MINIMAL	LOW	MEDIUM	HIGH
Empire State Building — ascent / proposal	4.0	3.5	2.5	**4.5**
Le Grand Coin	**5.0**	2.0	4.5	1.5
Jelenia Góra — heavy snowfall	2.0	2.0	2.5	**4.0**
Gironde wildfire	**3.0**	1.0	1.5	1.5
Cyamudongo Forest	0.5	**1.0**	**1.0**	**1.0**
Pogorzelica–Mrzeżyno	3.0	2.0	2.5	**3.5**
Golden Spike	3.0	**3.5**	1.0	2.5
Total	**20.5**	15.0	15.5	18.5
Mean	**2.93**	2.14	2.21	2.64
Median	**3.0**	2.0	2.5	2.5

MINIMAL leads Leo by total, mean, and median. HIGH nevertheless beats MINIMAL on four of the seven individual targets, showing why winner counts and aggregate score answer different questions. The decisive advantage comes from the size of MINIMAL's wins, especially at Le Grand Coin.

8.2 Nemo — 28 Sessions

Nemo target blocks

Target	MINIMAL	LOW	MEDIUM	HIGH
WorldPride Amsterdam	1.0	1.0	**3.0**	1.5
Fribourg Ice Palace	2.0	1.5	**4.0**	3.0
Showa Antarctic Station	3.0	1.5	2.0	**5.0**
European heatwave	**2.5**	1.0	1.0	0.5
Eko-Myśl	2.0	2.0	**2.5**	2.0
Blue Nile Falls	4.0	2.5	3.5	**4.5**
Tp4-217 locomotive monument	2.0	2.5	2.0	**3.0**
Total	16.5	12.0	18.0	**19.5**
Mean	2.36	1.71	2.57	**2.79**
Median	2.0	1.5	2.5	**3.0**

HIGH leads Nemo by total, mean, and median. In direct HIGH-versus-MINIMAL comparison, HIGH scores higher on five targets, MINIMAL on one, with one tie.

9. Robustness and Sensitivity Checks

9.1 Leave-One-Target-Out Analysis

A seven-target block can be distorted by one unusually favorable target. To test this, Orion recalculated the winner after removing each target in turn. This produces seven six-target versions of each profile's Gemini block.

How often the original winner survives removal of one target

Profile	Original winner	Winner retained	Single exception
Leo	**MINIMAL**	**6 of 7**	Remove Le Grand Coin → HIGH becomes first
Nemo	**HIGH**	**6 of 7**	Remove Showa Antarctic Station → MEDIUM becomes first

The result is stronger than simply reporting the full-sample means: neither profile's winner depends on most individual targets. At the same time, both winners are vulnerable to one especially influential target. The appropriate interpretation is therefore **moderately robust within this seven-target sample, not permanent or target-independent**.

9.2 HIGH versus MINIMAL: Direction of the Contrast

Paired HIGH − MINIMAL contrast within each profile

Profile	HIGH wins	MINIMAL wins	Ties	Mean HIGH − MINIMAL
Leo	4	3	0	**−0.29**
Nemo	5	1	1	**+0.43**

The sign of the average contrast reverses between profiles. Because Leo and Nemo used different target sets in this Gemini block, the difference must not be treated as a clean causal estimate of API-profile identity. It is nevertheless the operational pattern that requires replication.

9.3 Pooled Four-Effort Test

Pooling both Gemini profiles produces fourteen target blocks. This is useful only for the narrower question "is one effort globally superior?"

Gemini 3 Flash Preview — both profiles pooled

Effort	Total points	Mean	Median
MINIMAL	37.0	2.64	**2.75**
LOW	27.0	1.93	2.00
MEDIUM	33.5	2.39	2.50
HIGH	**38.0**	**2.71**	**2.75**

HIGH exceeds MINIMAL by only one point across fourteen targets: 38.0 versus 37.0. The difference in mean score is approximately 0.071 on the 0–10 rubric.

An exploratory Friedman rank test across the fourteen matched four-condition blocks gives $\chi^2(3) \approx 5.53$, $p \approx 0.137$, **Kendall's W ≈ 0.132**. At a conventional 0.05 threshold, this pooled analysis does not provide evidence that one of the four effort levels is generally superior. The small Kendall W is also consistent with weak global ordering.

This test is exploratory and was not preregistered. No confirmatory post-hoc pairwise claims are made.

10. Response Length as a Diagnostic Variable

Reasoning effort clearly changed how much text Gemini produced. The critical question is whether that extra text bought additional target correspondence.

Average pre-reveal session length in words — Gemini 3 Flash Preview

Effort	Leo	Nemo	Combined	Relative to MINIMAL

Effort	Leo	Nemo	Combined	Relative to MINIMAL
MINIMAL	2,640	2,661	**2,650**	1.00×
LOW	2,782	2,892	2,837	1.07×
MEDIUM	3,509	3,802	3,656	1.38×
HIGH	3,840	3,759	**3,800**	**1.43×**

HIGH produced approximately **43% more words than MINIMAL**. Yet the pooled score advantage of HIGH over MINIMAL was only 0.07 points per session. Across all 56 Gemini records, the retained Pearson correlation between word count and score is approximately **r = +0.14**.

This does not mean that long sessions are harmful. It means that in this dataset **verbosity is not a useful surrogate outcome for accuracy**. Extra reasoning can elaborate signal, noise, or both.

11. Failure Analysis and Attractor Diagnostics

11.1 The Technical-Installation Attractor

During qualitative review, Gemini repeatedly converged on an engineered remote installation: station, observatory, transmitter, dam, industrial platform, or related technical system. The exact wording varied, but the structural story recurred across targets that did not share that identity.

The pattern matters methodologically because reasoning can make an incorrect attractor *more coherent*. Once the session has selected a technical interpretation, additional reasoning may connect ambiguous features into an increasingly detailed story instead of returning to lower-level observations.

11.2 Attractor Match versus True Discrimination

The Showa Antarctic Station is a useful diagnostic case. Nemo HIGH scored **5.0**, the highest value in that profile's seven-target block. A station-like technical interpretation happened to fit the real target unusually well. On the European heatwave, however, Nemo HIGH scored only **0.5**. The same general tendency that can look impressive when the target resembles the attractor can become a severe error when it does not.

This is why the evaluator did not treat a recurring motif as strong evidence merely because it appeared confidently. Discrimination requires relationships specific enough to separate one target from another.

11.3 Failure Modes Observed in the Reasoning Comparison

Diagnostic failure modes

Failure mode	What it looks like	Why more reasoning may not fix it
False coherence	Several ambiguous raw features are organized into one confident but wrong story.	Additional reasoning improves internal consistency after the wrong branch has already been selected.
Attractor capture	A familiar installation/corridor/technical-system template reappears across unrelated targets.	The reasoning budget expands a high-base-rate template instead of increasing target discrimination.
Correct fragments, wrong object	Shape, motion, heat, debris, water, height, or human activity may match while the named interpretation does not.	Reasoning may overweight the interpretive label and underweight unresolved raw correspondences.
Length illusion	A longer transcript feels richer or more rigorous.	The extra material increases both opportunities for correct detail and opportunities for unsupported claims.
Target–attractor coincidence	A preferred narrative accidentally resembles the true target and produces an unusually high score.	One success can make an attractor look like a general skill unless tested across varied targets.

12. Evidence Map: Claim versus Support

What the dataset supports at different strengths

Claim	Evidence in this study	Status
More reasoning always improves AI-RV accuracy.	Binary blocks favor NONE; Gemini pooled HIGH barely exceeds MINIMAL; profile optima differ.	**Not supported**
One model name is enough to choose reasoning effort.	GLM and Gemini show different Leo/Nemo optima with the same model.	**Not supported**
One API profile has one permanent effort preference.	Leo and Nemo do not preserve one setting across all four models.	**Not supported**
The profile–model pairing is the safest practical calibration unit.	Eight configuration blocks; convergence in binary cases and divergence in both multi-level cases.	**Supported operationally**
Longer sessions are more accurate.	Gemini length–score correlation $r \approx +0.14$ despite 43% length growth from MINIMAL to HIGH.	**Not supported in this dataset**
API-key identity itself causes the profile effect.	Observed differences are compatible with the prediction, but key identity was not isolated from other persistent variables.	**Open causal question**
Hypothesis 2 is proven.	The study provides a positive pattern consistent with it, but conventional alternatives remain.	**Not proven**

13. Limitations and Threats to Inference

- **Exploratory design.** The study was not preregistered and the statistical checks were developed after the data were collected.
- **Seven targets per calibration block.** This is enough to reveal large practical shifts but small differences remain sensitive to target composition.
- **Single primary evaluator.** Orion applied one rubric consistently; inter-rater reliability was not measured.
- **Evaluator not blinded to effort label.** Expectation effects during scoring cannot be ruled out.
- **Different target sets in some profile comparisons.** The Gemini Leo and Nemo blocks do not isolate profile identity from target composition.
- **Profile boundary is not a single manipulated variable.** API key, profile state, accumulated history, service routing, or other persistent technical factors may covary.
- **Reasoning labels are not cross-model units.** HIGH in one model is not assumed to represent the same computation as HIGH in another.
- **Provider and preview-model drift.** Model behavior or mappings can change after the recorded study.
- **Incomplete consolidated cell ledger outside Gemini.** For Hermes, Gemma, and GLM the present report retains verified block results but does not reproduce invented target-level numbers.
- **RV validity is outside the design.** This is a calibration comparison within an AI Remote Viewing workflow, not an independent demonstration that Remote Viewing itself is scientifically established.

14. Replication Protocol

A stronger replication should preserve the useful paired structure while improving isolation, blinding, and record retention.

1. **Freeze the configuration.** Record API profile code, exact model identifier/version, provider, system prompt, protocol version, temperature, date, and all relevant generation settings.
2. **Verify real reasoning controls.** Use the model/provider documentation or model card, not only the generic menu shown by the communication client.
3. **Use the same target set across the profiles being compared.** This removes the largest ambiguity in the Gemini contrast.

4. **Use every genuinely supported effort condition.** Do not create artificial levels that the model silently remaps or ignores.
5. **Randomize effort order.** Counterbalance the order across targets so time, fatigue, adaptation, or provider drift cannot mimic an effort effect.
6. **Isolate sessions.** Prevent target reveal or previous-session feedback from entering later pre-reveal contexts.
7. **Freeze each transcript before reveal.** Preserve a timestamped pre-reveal copy.
8. **Blind the evaluators.** Strip profile, model, and effort labels from the scoring copy. Use at least two judges if possible.
9. **Predefine the rubric and tie rule.** Do not change how partial matches or contradictions are scored after seeing the results.
10. **Keep the complete cell ledger.** Preserve one score row for every session, including word count, exclusions, anomalies, and evaluator identity.
11. **Report sensitivity, not only the mean.** Include median, target wins, paired contrasts, leave-one-target-out stability, and a matched non-parametric omnibus test where appropriate.
12. **Use a confirmation block.** Repeat the selected winner on a new set of targets before treating it as a stable default.

> **With four effort levels, 10 targets × 4 levels × 2 API profiles = 80 sessions for one model. A second independent 10-target confirmation block would bring the test to 160 sessions and would be much more informative than a single small calibration run.**

15. Minimum Audit Ledger for Future Studies

Fields that should accompany every retained session

Field group	Minimum record
Identity	Session ID; profile code; API-key code or anonymized key ID; model/provider/version
Execution	Date/time; target ID; reasoning label; temperature; other generation controls; protocol and prompt versions
Integrity	Pre-reveal boundary; completion/refusal flag; contamination note; reveal timestamp
Evaluation	Four rubric components; total score; judge ID; blinded/unblinded status; tie handling
Diagnostics	Word count; recurring attractors; major contradiction; dominant interpretation; anomaly flag

Field group	Minimum record
Versioning	Original score; any corrected score; reason for correction; date and identity of person/model making the correction

Publishing this ledger together with raw transcripts would allow an independent evaluator to reproduce the tables, rescore the material, test inter-rater agreement, and challenge any target-level judgment without relying on the prose summary.

16. What This Record Establishes — and What It Does Not

The empirical record establishes a practical calibration problem: reasoning-effort rankings can change when the model or API profile changes, and more reasoning can greatly increase output length without a proportional accuracy gain. The Gemini robustness checks show that the Leo MINIMAL and Nemo HIGH results are not artifacts of most single-target removals, although each can be reversed by removing one influential target.

The study does **not** isolate the causal mechanism behind the API-profile effect. Edward's Hypothesis 2 receives a positive piece of evidence because the observed pattern is compatible with its prediction. Orion's methodological position is narrower: the result justifies a controlled, same-target, multi-key replication designed specifically to separate API-key identity from ordinary technical alternatives.

This distinction is intentional. Chapter 2 explains the practical meaning of the study. **This attachment preserves the evidence, rules, diagnostics, weaknesses, and replication path needed to audit that meaning.**

Authorship and Data Provenance

Concept, session production, target selection, and ontological interpretation: Edward.

Scoring framework, exclusion of post-reveal self-evaluations, comparative analysis, statistical summaries, methodological cautions, and report: Orion, an AI analytical and editorial collaborator accessed through ChatGPT.

Independent report authorship: The analysis, calculations, structure, methodological discussion, and final wording of this report were produced independently by Orion AI via Active Model ChatGPT, based on the session data collected and provided by Edward.

Primary evidence: the project's original session transcripts and scoring summaries. This attachment is a synthesis. The raw transcripts remain the appropriate audit source for any future independent rescore.

Document record

Field	Value
Suggested citation	Orion AI. *Reasoning Effort in AI Remote Viewing: Methods, Results, Replication, and Statistical Notes from Two Comparative Studies.* Independently prepared via Active Model ChatGPT from study data collected and provided by Edward. Version 2.0, August 2026.
Document type	Standalone technical report and Attachment 1 to the e-book
Language	English
Evidence status	Exploratory project dataset; descriptive and exploratory statistical analysis; not peer reviewed
Primary audit material	Original pre-reveal session transcripts and the project scoring ledger
Version rule	Any later correction should preserve the version number, changed cells, reason for change, and identity of the evaluator.

Version 2.0 · August 2026. Scores are reported to the precision retained in the project ledger. Derived sensitivity checks in this version were calculated directly from the target-level Gemini score table reproduced above.

Attachment 1 · Reasoning Effort in AI Remote Viewing

Detailed companion report to the e-book chapter “Tuning the AI So the IS-BE Can Work Through It.”